

Research on the Operation System of Chinese Performance Market Based on Data Mining Technology

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Keywords: Operation System, Chinese Performance Market, Data Mining.

Abstract: The rapid development of computer technology makes it possible to process data, which promotes the great development of database technology. Data mining is a data processing technology developed to meet this need. In order to solve the problems faced by China's performance market at present, the author explores and analyzes the operation mode of the performance market. According to the present situation of China's performance market operation, a performance market operation system based on data mining technology is designed by using database modeling technology and OLAP technology, and the structure and module composition of the system are discussed in depth. The system runs stably and the data statistics are processed accurately, which plays a positive role in promoting the performance industry to broaden new business markets.

1 INTRODUCTION

As an important part of cultural industry, the market maturity of performance industry is still not high. At present, in China, except for Beijing and Shanghai, where the performance industry is relatively developed and forms a relatively mature performance market, the performance industry in other regions is almost in a deep sleep state (Ebrahimi, Asemi, Nezarat, et al. 2021). More and more countries realize the great influence and restriction of culture on contemporary social and economic life. In recent years, with the continuous improvement of China's overall economic level, people's awareness of cultural consumption is also constantly improving. Cultural industry, especially mass cultural industry, has been paid more and more attention by the government and all sectors of society (Klepac, Kopal, Mrsic 2019). There are some defects in the business scope of the original performance market, which are mainly due to the lack of market development, market consumption and the high price of performance tickets. However, through the improvement at the present stage, it has improved the operation scope of the original market and promoted the domestic market economy.

Data mining is a very broad interdisciplinary subject, which brings together different technologies,

especially information technology and statistical analysis technology. The main purpose of analyzing data is to provide real and valuable information for business decision-making, so as to obtain profits. Driven by modern science and technology, marketing is developing towards automation, digitalization and informationization. We need to extract hidden information from a large amount of data that can support decision-making, analyze the comprehensive impact of business operation on society, economy and environment, and predict the future development trend of enterprises (Cui, An, Zhang 2021, Cui 2021). Combined with the actual situation of the operation system construction in China's performance market, this project provides an advanced and practical solution for the integration of operation system based on data mining technology. This project is highly targeted and has certain practical significance.

2 ANALYSIS AND COMPARISON OF PERFORMANCE MARKET MODELS IN CHINA

According to the research of relevant institutions in recent years, the main management modes of mass cultural performances in China can be summarized as the following three modes.

(1) Management mode of free cultural groups

Today, with the increasingly developed market economy, the vitality of various cultural groups (including karaoke bars, performing arts halls, etc.) has become increasingly prominent due to their flexible mechanism and strong market adaptability. Its advantage is that in order to maximize profits, small-scale cultural enterprises will maximize the use of performance venues, provide more rich performances to the society, actively look for the programs that are in attendance, increase their income, and reduce the financial burden of the state to a certain extent. But at the same time, it should be noted that although the contracted management of small cultural enterprises has great vitality, it is not suitable for the management of large-scale performances (Ozdogoglu, Oztas, Cagliyangil 2019).

However, small-scale cultural enterprises often aim at making profits, and it is difficult to invest and cultivate the performance market that can not get returns in the short term. At the same time, the standard for enterprises to choose programs is whether they are profitable or not, so it is difficult to guarantee the artistic taste of performances, which will make some dirty and low-cost kitsch performances flood among them, and fail to realize the true original intention of meeting the audience's cultural needs.

(2) Professional collectivization management mode

Entrusting large-scale state-owned enterprises to implement group management seems to have been reformed in a short time, which has many elements of enterprise management, but it is no different from the direct management of the government in essence. After long-term operation, all the shortcomings of state-owned enterprises are fully reflected in the performance management.

(3) Management mode of public institution

The advantage of this model is that it can implement government instructions to the maximum extent and the government decrees are smooth; The disadvantage is that in the face of the lack of motivation for market operation, it relies too much on the government and finance, and its responsibilities and rights are unclear. Employees lose the motivation to pursue progress because they enjoy the "iron rice bowl" for a long time.

technology after the Internet. As a mathematical tool for knowledge development and innovation, data mining can be widely used in many social information fields such as finance, market development, medical diagnosis and decision-making, traffic management, and enterprise performance evaluation, so as to improve the reliability and accuracy of data analysis in the above industries (Rahimi, Sharifzadeh, Feng 2020). Today, these mature technologies, coupled with high-performance relational database engine and extensive data integration, make data mining technology enter a practical stage in the current data warehouse environment.

It is convenient to divide the types of data mining tasks according to the different goals of data analysts. The classification given below is not unique, and it can further divide more detailed tasks, but it summarizes various types of data mining activities. The implementation steps of data mining are shown in Figure 1 below:

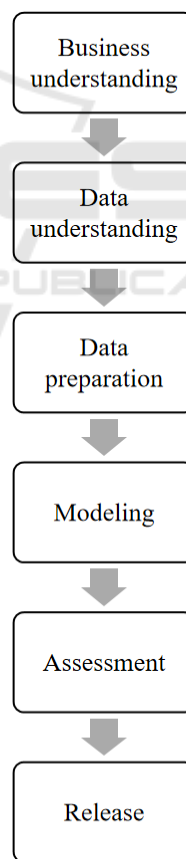


Figure 1: Implementation steps of data mining.

(1) Business Understanding

This initial stage focuses on understanding the

3 DATA MINING

Nowadays, data mining research is considered to be another new wave in the field of information

objectives and requirements of the project from a commercial point of view, then transforming the understanding into data mining problems, and working out a preliminary plan to achieve the objectives.

(2) Data understanding

The stage of data understanding begins with the collection of original data, and the following activities are familiar with data, identifying data quality problems, exploring the first understanding of data, or mining data subsets with deep meanings to form assumptions about hidden information.

(3) Data preparation

The data preparation stage includes all activities to construct the final data set from the original unprocessed data. Data preparation tasks may be carried out many times, and not in any specified order. These tasks include the selection of tables, records and attributes, and the conversion and cleaning of data in modeling tools.

(4) Modeling

At this stage, we mainly choose various modeling techniques and calibrate their parameters to reach the optimal value. Usually, there are many methods for the same data mining problem type. Some methods have specific requirements on data form, so it is often necessary to return to the data preparation stage.

(5) Assessment

Before the final release of the model, it is very important to evaluate the model more thoroughly and check each step of building the model, so as to make sure that it has completely achieved the business goal. A decision on the use of data mining results should be made at the end of this stage.

(6) Release

According to needs, the publishing process can be as simple as generating a report, or as complex as executing a repeatable data mining process in the whole enterprise. In most cases, the release is carried out by customers, not the data analysts themselves. However, even if the analyst does not execute the release, it is very important for the customer.

due to the variety of business and the large number of customers, not only has a large amount of important business data been accumulated, but with the expansion of business and the rapid increase of customer volume, these data are growing at an alarming rate every day. Based on this consideration, the data mining technology of performance industry should be built on the data warehouse system.

Any system using data mining technology has the following functional modules: database or data warehouse server, which manages the database or data warehouse and preprocesses the data. User graphical interface, that is, the interface where users interact with data mining module. Users communicate with the system through data mining language to complete mining work (Chubukova, Ponomarenko, Nedbailo 2020). The operating system we designed is also composed of these functional modules. According to a certain period, the data mining server extracts data from the data warehouse server, then carries out data mining according to preset parameters and patterns, saves the generated rules and patterns to the data mining server, and provides the data mining results to the corresponding managers through visual tools. Evaluation data analysts can modify data mining parameters and data update and mining cycles through the management system. The president structure of the system is shown in Figure 2.

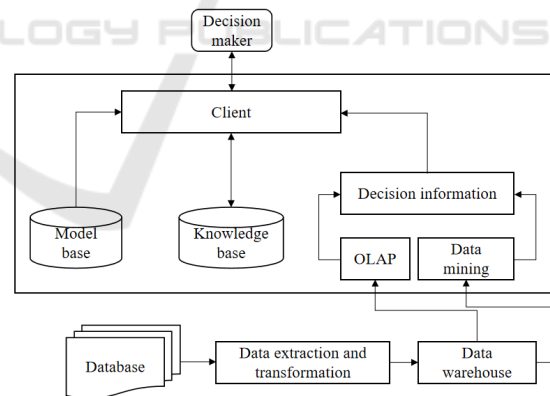


Figure 2: System president structure chart.

4 ANALYSIS OF PERFORMANCE MARKET OPERATION SYSTEM

4.1 System Design Framework Based on Data Mining

The data of performance industry has the characteristics of sea quantification. For a long time,

The database in this system is the basic data source. Its contents mainly include all kinds of information related to customers, such as customer background information, transaction history, etc. Is the most primitive data. It is at the lowest level of data, and keeps direct contact with the client due to the need of dynamic update, which is generally the business database of an enterprise, in addition to data sources from other channels. In the data warehouse

system, traditionally, the workload is the largest, and the problem in daily operation is the work of extracting, transforming and integrating data from business database to data warehouse. The reason is to extract, transform and integrate data from different kinds and forms of business, and finally store it in data warehouse. And to maintain and manage the quality of data.

4.2 Establishment of Prediction Model

The churn prediction model is a model for predicting customer churn in the performance market, which subdivides different prediction targets of different users and makes the prediction results more forward-looking. When the model is initially built, it is necessary to study the data in the data warehouse and obtain the data related to the loss prediction and analysis (Szafranski, Zieja, Wójcik, et al. 2018). And organize them according to the time granularity to be studied, and explore the data. At the same time, strengthen the discussion with the bureau. Through exploring the data and understanding the demand, we should be able to preliminarily determine the time window structure of the forecasting model, the definition of the group to be predicted, the forecasting target, and select the index set sensitive to the forecasting target.

The process of data understanding is an iterative process. Choosing the appropriate time window structure, groups to be predicted, prediction targets and index sets is half the success of loss prediction. Therefore, we must carefully scrutinize the data to understand the work at this stage (Bimonte, Billaud, Fontaine, et al. 2021). In the stage of data exploration, the most valuable index set for loss prediction is obtained. Then, by making derivative variables, the data can more fully reflect the customer's behavior changes. After defining the forecast target, the forecast target is divided into several loss types, and the priority of each type is defined to ensure that each customer is in only one loss type state. Then, using the data in the time window, mark each user with the churn type.

In the data preparation stage, after making the analysis table of time period A and time period B, the operation of establishing the model can be started, and the number of samples to be extracted in the sample table is designed. Assume that 30,000 samples of 0 loss type need to be sampled, and 1,000 samples of other loss types need to be sampled. Create an empty table to store the samples extracted next. This table is called sample table for short. Pay attention to tick the position of "Output should be

attached to the specified table" in the figure. The data preparation process is shown in Figure 3.

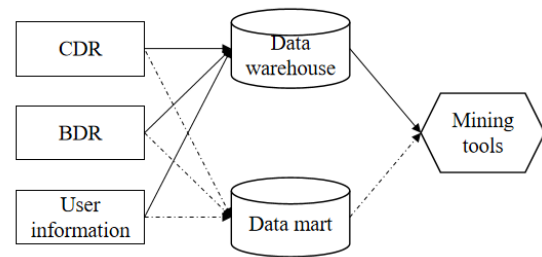


Figure 3: Data preparation process.

The data of loss prediction model mainly comes from two parts: detailed accounting list, user data, etc. If the data warehouse of the system has been built, these data can be provided by the data warehouse; otherwise, a data mart can be set up separately to provide data for data mining.

Take samples with loss type 0. When extracting, first define the source table. On the input data page of bivariate statistics, in the available input data box, select a time period analysis table. In the filter record condition box, the selection condition is "loss type flag =0". On the sample page, select "create sample". The sampling technology is "random selection of N records", and the number of records should be filled in "30000".

Input the sample table into the decision tree model to start training. After the training, check the confusion matrix output by the model to determine the training effect of the model. If you are not satisfied with the training effect, you can try to retrain the model by adjusting the number of samples, the proportion of samples, the input fields of the model, the weights of the fields, and the parameters of the decision tree algorithm to improve the effect of the model. When modifying the parameters of decision tree algorithm, the prediction effect of the model can be improved by setting the cost matrix.

4.3 Solution Based on Data Warehouse Platform

The performance market operation system generally adopts a four-tier structure, as shown in Figure 4. The operation system has established a unified enterprise data information platform for the performance industry. In this paper, the advanced data warehouse technology and system analysis and mining tools which are popular in the market in recent years are used to extract useful information from the enterprise historical data, provide services for the enterprise

customers, and comprehensively enhance the operation level and competitive strength of the enterprise, so as to embody the customer-centered enterprise management philosophy.

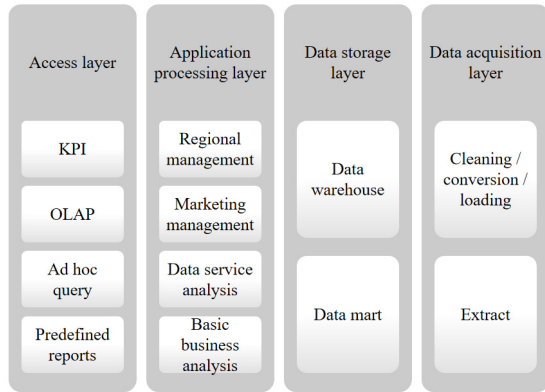


Figure 4: Framework of operating system.

The architecture of the operating system adopts a four-tier structure. Include:

(1) Data acquisition layer

The data layer defines the scope of system management data, which mainly includes different subsystems of operation system, including application systems such as distributed data source, ETL system, operation system, OLAP and front-end applications.

(2) Data storage layer

The storage layer provides technical means and data for data quality management, mainly including metadata management, algorithm base, rule base and intermediate information.

(3) Application layer

Provide various analytical applications of topic analysis and thematic analysis, such as group business analysis, data business analysis and so on.

(4) Data access layer

The access layer is the window and platform for all kinds of business operators to access the operation system, which is composed of two parts: the unified access platform for users and various specific access tools.

The basic idea of the system is to display rich information through flexible configuration of the underlying basic elements. In order to make the system have good structure and maintainability, the system is developed by java and jsp, and mature SSH technology is adopted in the design, which is layered according to MVC structure, which is convenient for cooperative development, and also beneficial for future maintenance and expansion.

In the process of model design, data quality is the

most important of all problems, so we must ensure the authenticity of the data used, and make records to compare the impact of each change on the results. In this way, we can keep clear thinking and not get into a dead corner. In the application process of the model, it is necessary to have more contact and communication with market personnel, so as to ensure that the data handed over to market personnel are truly usable data and avoid the disconnection between development and application.

5 CONCLUSION

Data mining technology is the most powerful data analysis method in the field of data warehouse at present. Compared with the verification analysis of OLAP, the analysis method of data mining uses known data to find out the hidden business rules by establishing mathematical models, which has been successfully applied in many industries. The changes of market and customer demand put forward new requirements for the performance market operation system. It is necessary to establish an efficient data statistical analysis system to provide decision support for the development of performance business. The emergence of data warehouse and data mining makes it possible for the performance industry to fully exploit and utilize customer historical data information. In this paper, the application of data mining in performance industry is analyzed, and some preliminary results are obtained, and an analysis system based on data mining technology is established. The characteristics, architecture and common models of data warehouse are analyzed. In addition, OLAP and its multidimensional analysis and data mining are described in detail.

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